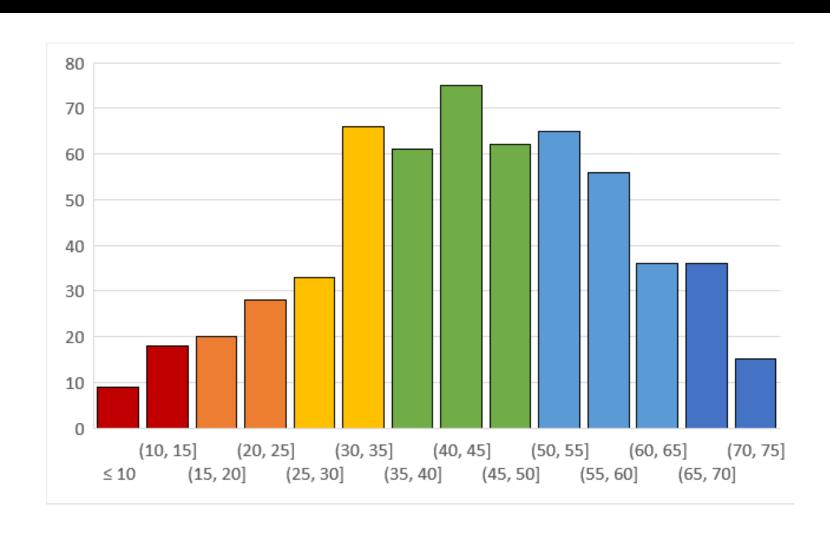
CS1010S Programming Methodology

Lecture 9 Generic Operations

24 October 2018

Midterm Results



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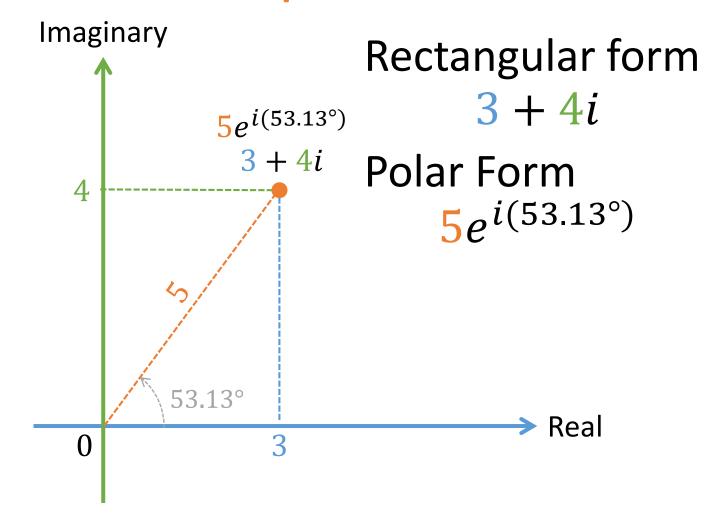
Course Hero

Agenda

- Recap: Complex-Arithmetic Package
- Tagging Data
- Implementing Generic Operators
 - Dispatch on Type
 - Data-Directed Programming
 - Message Passing

Last Week Multiple Representations

Recall: Complex Numbers



Abstraction barrier

Programs that use complex numbers

add_complex, sub_complex, mul_complex, div_complex

Complex Numbers Package

Rectangular representation

Polar representation

Arithmetic package

Similar to rational number package (Lecture 7)

```
Addition: z = z1 + z2

real_part(z) = real_part(z1) + real_part(z2)

imag_part(z) = imag_part(z1) + imag_part(z2)
```

Arithmetic package

```
magnitude(z) = magnitude(z1) * magnitude(z2)
angle(z) = angle(z1) + angle(z2)
etc.
These are easily implemented assuming the existence of
selectors: real_part, imag_part, magnitude, angle
constructors: make from real imag, make-from-mag-ang
```

Multiplication: z = z1 * z2

Rachel's (rectangular) code

```
def make_from_real_imag(x, y):
    return (x, y)
def real part(z):
    return z[0]
def imag part(z):
    return z[1]
def magnitude(z):
    return math.hypot(imag part(z), real part(z))
def angle(z):
    return math.atan(imag part(z) / real part(z))
def make_from_mag_ang(r, a):
    return (r * math.cos(a), r * math.sin(a))
```

Code in Action

```
def print complex(z):
    print(str(real part(z)) + " + " str(imag part(z)) + "i")
a = make_from_real_imag(1, 1)
b = make from real imag(2, 2)
print complex(a)
                   \rightarrow 1 + 1i
print complex(b) \rightarrow 2 + 2i
def add complex(z1, z2):
    return make from real imag(real part(z1) + real part(z2),
                                 imag part(z1) + imag part(z2))
print complex(add complex(a, b))
                                      \rightarrow 3 + 3i
```

Rachel's (polar) code

```
def make_from_mag_ang(r, a) :
    return (r, a)
def real part(z):
    return magnitude(z) * math.cos(angle (z))
def imag part(z):
    return magnitude(z) * math.sin(angle(z))
def magnitude(z):
    return z[0]
def angle(z):
    return z[1]
def make_from_real_imag(x, y):
    return (math.hypot(x, y), math.atan(y / x))
```

Same Code!!

```
a = make from real imag(1, 1)
b = make from_real_imag(2, 2)
def add complex(z1, z2):
  return make from real imag(real part(z1) + real part(z2),
                imag part(z1) + imag_part(z2))
```

Two Lessons

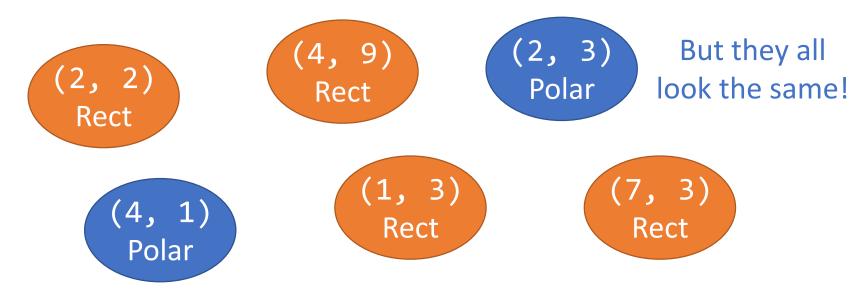
- Different representations might have slightly different accuracy because of internal representation
- Power of data abstraction
 - Can reuse same code even though the representation is completely different

Co-existing Multiple Representations:

What are the issues?

```
Rect version:
make_from_real_imag(x, y)
real_part(z)
imag_part(z)
magnitude(z)
angle(z)
make_from_mag_ang(r, a)
```

```
Polar version:
make_from_real_imag(x, y)
real_part(z)
imag_part(z)
magnitude(z)
angle(z)
make_from_mag_ang(r, a)
```



All tuples!!

Co-existing Multiple Representations

1. Matching representations to operations.

- e.g. rectangular rep. must be given to rectangular functions.
- Polar rep. cannot be given to rectangular functions, or vice versa.

2. Name conflicts

- Both reps have functions with identical names!
- e.g. real_part, angle, etc.

Matching rep. to operations

Solution: tagged data

- Each representation is given a tag to explicitly indicate the representation (type).
- e.g. rectangular (<rectangular data>)
- e.g. polar (<polar data>)
- Use tuple: (tag, content)

Tagging

```
def attach tag(type tag, contents):
    return (type tag, contents)
def type tag(datum):
    if type(datum) == tuple and
       len(datum) == 2:
        return datum[0]
    else:
        raise Exception('Bad tagged datum --type tag' +
str(datum))
                  Exception: Signals something
                      bad has happened
```

Tagging

```
def contents(datum):
    if type(datum) == tuple
        and len(datum) == 2:
        return datum[1]
    else:
        raise Exception('Bad tag datum -- contents ' +
str(datum))
```

Matching rep. to operations

Rachel's revised constructor:

```
def make_from_real_imag(x, y):
    return attach_tag('rectangular', (x, y))
```

Tagged data

Rachel's revised constructor:

Tagged data

To check for tag:

```
def is_rectangular(z):
    return type_tag(z) == 'rectangular'

def is_polar(z):
    return type_tag(z) == 'polar'
```

Name conflicts

But both Rachels' functions still have the same name!

Whoops!

Name conflicts

Solution: impose naming convention function name ends with tag

Have we solved the problem?

Whose responsibility?

- Sanity check: whose responsibility is it to match data type with the corresponding operations?
 - User?
 - Rachel?
 - Rachel?
 - Other?

Let's Make it the User's Problem?

Programs that use complex numbers

add_complex, sub_complex, mul_complex, div_complex

Complex Numbers Package

Rectangular representation

Polar representation

User's responsibility

User needs to check for rep. and call appropriate functions:

```
def add complex(z1, z2):
    if is rectangular(z1):
        real1 = real part rectangular(contents(z1))
        imag1 = imag part rectangular(contents(z1))
    else:
        real1 = real part polar(contents(z1))
        imag1 = imag_part_polar(contents(z1))
    if is rectangular(z2):
        real2 = real part rectangular(contents(z2))
        image2 = imag part rectangular(contents(z2))
    else:
        real2 = real part polar(contents(z2))
        imag2 = imag_part_polar(contents(z2))
    return make from real imag rectangular(real1 + real2, imag1 + imag2)
```

A lot of code devoted to checking!

Rachel's (rectangular) code

```
def make_from_real_imag_rectangular (x, y):
    return (x, y)
def real_part_rectangular(z):
    return z[0]
def imag part rectangular (z):
    return z[1]
def magnitude rectangular (z):
    return math.hypot(imag_part(z), real_part(z))
def angle_rectangular (z):
    return math.atan(imag part(z) / real part(z))
def make_from_mag_ang_rectangular (r, a):
    return (r * math.cos(a), r * math.sin(a))
```

Rachel's (polar) code

```
def make_from_mag_ang_polar(r, a) :
    return (r, a)
def real part polar(z):
    return magnitude(z) * math.cos(angle(z))
def imag part polar(z):
    return magnitude(z) * math.sin(angle(z))
def magnitude polar(z):
    return z[0]
def angle_polar(z):
    return z[1]
def make_from_real_imag_polar(x, y):
    return (math.hypot(x, y), math.atan(y / x))
```

Whither the future?

- One or more of the following may happen in the future:
 - Rectangular code is removed. Only polar representation is left.
 - A new representation for complex numbers is installed, and co-exists with current representations.
 - Rachel provides a new function for her rep. that no one else can provide.
- Who should now be responsible, so as to minimize the effects of the above changes?

MANAGING COMPLEXITY

Generic Operators

- Create another layer of abstraction.
 - To provide generic operators.
 - To shield user from the complexity of managing multiple representations.
- Rachel, Rachel or user can create this layer. Or the project leader.

Abstraction barrier

Programs that use complex numbers

add_complex, sub_complex, mul_complex, div_complex

Complex Numbers Package

Rectangular representation

Polar representation

New Abstraction Layer

Programs that use complex numbers

add_complex, sub_complex, mul_complex, div_complex

Complex Numbers Package

Generic Operators

Rectangular representation

Polar representation

Add new layer

Generic operators

User sees this generic operator.

```
def real_part(z):
    if is_rectangular(z):
        return real_part_rectangular(contents(z))
    elif is_polar(z):
        return real_part_polar(contents(z))
    else:
        raise Exception('Unknown type -- real_part' + z)
```

Type checking, tag removal, are hidden from user.

Generic operators

• User's code:

- Greatly simplified!
- Selectors are generic operators.
- Constructor: we can choose the best one
 - Use rectangular when we have real-imag.
 - Use polar when we have magnitude-ang.
- How do we implement this?

Strategy #1: Dispatching on type

- This strategy of providing generic operators based on checking the data type and calling the appropriate function is called *dispatching on type*.
- Generic operators work on data that could take on *multiple forms* (*polymorphic data*).

- 1. User calls generic operator to work on polymorphic data.
- Generic operator dispatches on type. Data is stripped of tag going down.
- 3. Returned data is tagged with type going up.

Programs that use complex numbers

add_complex, sub_complex, mul_complex, div_complex

Complex Numbers Package

Generic Operators

Rectangular
representation

2 Polar
representation

Moving across barriers

Issues to think about

- What are the advantages of providing generic operators?
 - Benefits to the user?
 - Impact of future code changes?
- What are the pros/cons of dispatching on type.
 - What changes are needed when a new representation is installed?
 - Does it resolve the name conflict?

Problems with Dispatch on Type

- Generic operators need to know all the types available.
- Adding a new type means changing all operators to dispatch correctly.
- Does not resolve name conflict.

Strategy #2: Data-directed programming

- How about using a table and doing a table lookup?
- Generic operator look at tag on data and find the correct operation from table
- Address problem of naming conflicts
- Allows easy extension: just add more entries to the table!

Table of operations

Type

		Polar	Rectangular
Operations	real_part	real_part	real_part
	imag_part	imag_part	imag_part
	magnitude	magnitude	magnitude
	angle	angle	angle

In creating generic operators, we were really selecting the appropriate lower-level function based on operation and type, as summarized in table above.

Table manipulation

- Suppose we have the following:
 - put(<op>, <type>, <item>)
 installs <item> in table, indexed by <op> and <type>
 - get(<op>, <type>)
 looks up <op>, <type> entry in table and returns item found there.

Installing rect package

```
def install_rectangular_package():
   # internal functions
   # unchanged from Rachel's initial code
  def make from real imag(x, y):
      return attach_tag('rectangular',(x, y))
  def real part(z):
      return z[1]
  def imag_part(z):
      return z[2]
  def magnitude(z):
      return math.hypot(real part(z),imag part(z))
  def angle(z):
      return math.atan(imag_part(z)/ real_part(z))
  def make from_mag_ang(r, a)
      return make from real imag (r * math.cos(a), r *
  math.sin(a))
```

Installing rect package

```
# cont'd
# interface to the rest of the system
def tag(x):
    return attach tag('rectangular', x)
put('real part', ('rectangular', ), real part)
put('imag part', ('rectangular', ), imag part)
put('magnitude', ('rectangular', ), magnitude)
put('angle', ('rectangular', ), angle)
put('make from real imag', 'rectangular',
     make from real imag)
put('make_from_mag_ang', 'rectangular',
     make_from_mag_ang)
return 'done'
```

Installing polar package

```
def install polar package():
  # internal functions
  # unchanged from Rachel's original code
  def make_from_mag_ang(r, a):
      return attach tag('polar', (r, a))
  def magnitude(z):
         return z[1]
  def angle(z):
      return z[2]
  def real part(z):
      return magnitude(z) * math.cos(angle(z))
  def imag_part(z):
       return magnitude(z) * math.sin(angle(z))
  def make_from_real_imag(x, y):
       return make_from_mag_ang(math.hypot(x, y),
                                 math.atan(y / x))
```

Installing polar package

Observation

- No name conflicts!
 - Because all function names are internal (local) to the installer.
 - Thus, each programmer can use identical names.
 - Names live in separate name space.
- Installer defines all operators and places them in the table according to operation and type.
- Tags still required to distinguish between types.

Table manipulation

- Suppose we have the following:
 - put(<op>, <type>, <item>)
 installs <item> in table, indexed by <op> and <type>
 - get(<op>, <type>)
 looks up <op>, <type> entry in table and returns item found there.

Deal with this assumption

- Often convenient to have a data structure that allow retrieval by keyword, i.e. put + get
- Table of key-value pairs. Commonly called Associative arrays
- Python dictionaries use the curly braces {}

```
{key1:value1, key2:value2, ...}
>>> {} # empty dict
>>> weather = { 'temp':{2:26.8, 14:31.1},
                'wind':0, 'desc':'cloudy'}
                     is equivalent to
>>> weather = dict(temp={2:26.8, 14:31.1},
                   wind=0, desc='cloudy')
```

more dictionary constructors: >>> dict([(1, 2), (2, 4), (3, 6)]){1:2, 2:4, 3:6} >>> weather = { 'temp':{2:26.8, 14:31.1}, 'wind':0, 'desc':'cloudy'} >>> weather['temp'] Accessed using {2:26.8, 14:31.1} key >>> weather['tomorrow'] KeyError: 'tomorrow'

```
>>> 'wind' in weather Checks if
                           key exists
True
>>> 0 in weather
False
>>> weather['is nice'] = True # adds an entry
>>> del weather['temp'] # delete an entry
>>> list(weather.keys())
['desc', 'is nice', 'wind']
>>> list(weather.values())
['cloudy', True, 0]
```

Looping construct

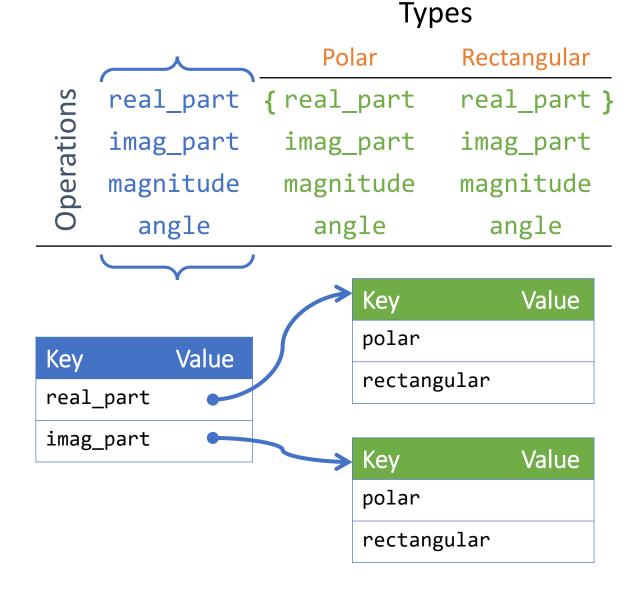
```
>>> for key in weather:
       print(weather[key])
cloudy
True
0
>>> for key, value in weather.items():
        print(key, value)
desc cloudy
is nice True
wind 0
>>> weather.clear() # delete all entries
```

Table implementation

```
def put(op, type, proc):
    if op not in procs:
        procs[op] = {}
    procs[op][type] = proc
```

```
def get(op, type):
    return procs[op][type]
```

2D associative array



Resulting Table

function name

```
>>> procs
                                  representation
{'real_part':
     {('polar',): <function install_polar_package.<locals>.real_part at xxx,
      ('rectangular',): <function install_rectangular_package.<locals>.real part at xxx},
 'imag_part':
     {('polar',): <function install_polar_package.<locals>.imag_part at xxx>,
      ('rectangular',): <function install_rectangular_package.<locals>.imag_part at xxx},
 'magnitude':
     {('polar',): <function install_polar_package.<locals>.magnitude at xxx,
      ('rectangular',): <function install_rectangular_package.<locals>.magnitude at xxx},
 'angle':
     {('polar',): <function install_polar_package.<locals>.angle at xxx,
      ('rectangular',): <function install_rectangular_package.<locals>.angle at xxx},
```

Making generic operators

```
def apply_generic(op,(*args));
    type_tags = tuple(map(type_tag, args))
    proc = get(op, type_tags)
    return pro((*args))
def type_tag(datum):
    if type(datum) == tuple
      and len(datum) == 2:
             return datum[0]
        else:
          raise Exception('Bad tagged datum --
                            type tag' + datum)
```

The * notation

For variable number of arguments

```
def f(x, y, *z):
     <body>
```

- function f can be called with 2, or more arguments.
- Calling f(1, 2, 3, 4, 5, 6): in the body,
 x → 1,
 y → 2,
 z → (3, 4, 5, 6)

The * notation

 You can also call a function for which you don't know in advance how many arguments there will be using *.

```
def funky(op, args):
    return op(*args)
print(funky(lambda x: x*x, (2,))) → 4
print(funky(lambda x,y: x+y, (2, 1))) → 3
```

Making generic operators

```
def apply generic(op, *args):
    type_tags = tuple(map(type_tag, args))
    proc = get(op, type tags)
    return proc(*args)
Example:
apply_generic('real_part', z1)
    op → 'real_part'
    args \rightarrow (z1)
    type_tags → ('polar',)
    proc → get('real_part', ('polar',) → lambda...
    lambda...(z1)
```

Making generic operators

```
def apply generic(op, *args):
    type_tags = tuple(map(type_tag, args))
    proc = get(op, type tags)
    return proc(*args)
def real part(z):
        return apply_generic('real_part', z)
def imag part(z):
       return apply_generic('imag_part', z)
def magnitude(z):
       return apply generic('magnitude', z)
def angle(z):
       return apply generic('angle', z)
```

Note:

 apply_generic retrieves appropriate function proc from table based on <op>,<type>.

Then applies proc to list of arguments.

 This style of using a table to dispatch is called datadirected programming.

Issues to think about

What advantages does data-directed programming have over dispatching on type?

- Consider the effect of adding a new type or new operation. Any code changes?
- Any name conflicts?
- Who should maintain table?

Generic Arithmetic

- We can push this idea further!
- We can use generic operators to handle completely different data, not just different representations of the same data.
- Let's build a generic arithmetic package.
 - Works with rational numbers, complex numbers, ordinary numbers, even polynomials!
 - Provide generic operators: add, sub, mul, div

Complex Numbers

- 1. User calls generic operator to work on polymorphic data.
- Generic operator dispatches on type. Data is stripped of tag going down.
- 3. Returned data is tagged with type going up.

Programs that use complex numbers

Complex Numbers Package

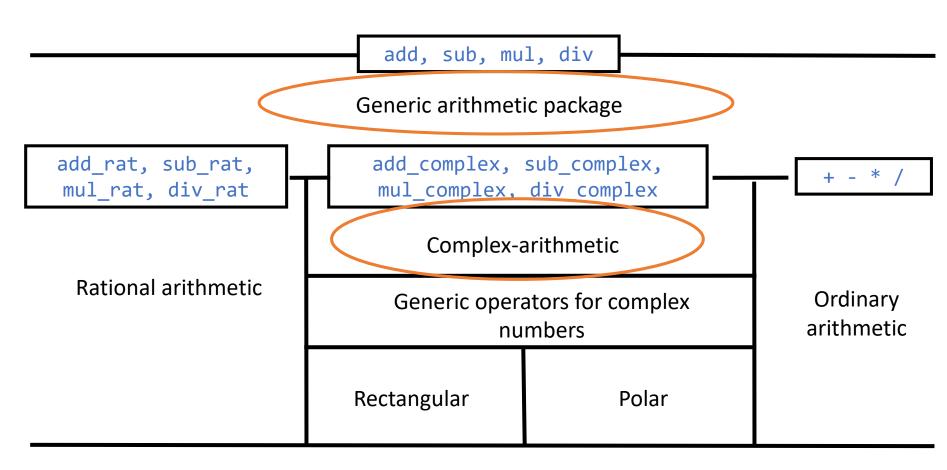
1 Generic Operators

Rectangular representation

A polar representation

Generic Arithmetic

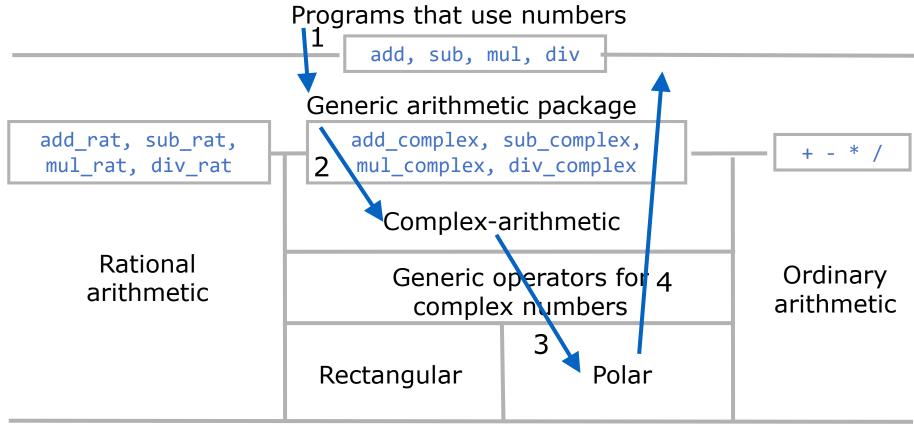
Programs that use numbers



Lists and primitive machine arithmetic

Generic Arithmetic

- 1. User calls generic arithmetic operators to work on polymorphic numbers.
- Generic operator dispatches on type. Data is stripped of tag going down, e.g. complex
- 3. Complex package calls generic operators for rect/polar forms. Tag is stripped for data going down.
- 4. Returned data tagged with types going up: polar, and complex.



Lists and primitive machine arithmetic

Implementation strategy

Use data-directed programming, rather than dispatch on type

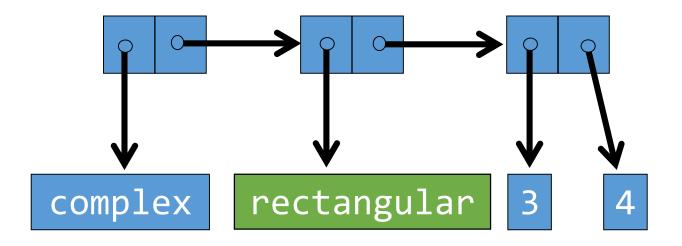
Generic Arithmetic

```
def install_complex_package():
 # imported functions from rectangular and polar packages
 def make from real imag(x, y):
     return get('make from real imag', 'rectangular')(x, y)
 def make from mag ang(r, a):
     return get('make_from_mag_ang', 'polar')(r, a)
 # internal functions
 def add_complex(z1, z2):
     return make_from_real_imag(real_part(z1) + real_part(z2),
                           imag part(z1) + imag part(z2))
 def sub complex(z1, z2):
         return make from real imag(real part(z1) - real part(z2),
                           imag part(z1) - imag part(z2))
 def mul complex(z1, z2):
    return make from mag ang(magnitude(z1) * magnitude(z2),
                       angle(z1) + angle(z2))
 def div complex(z1, z2):
     return make_from_mag_ang(magnitude(z1) / magnitude(z2),
                       angle(z1) - angle(z2)
```

Generic Arithmetic

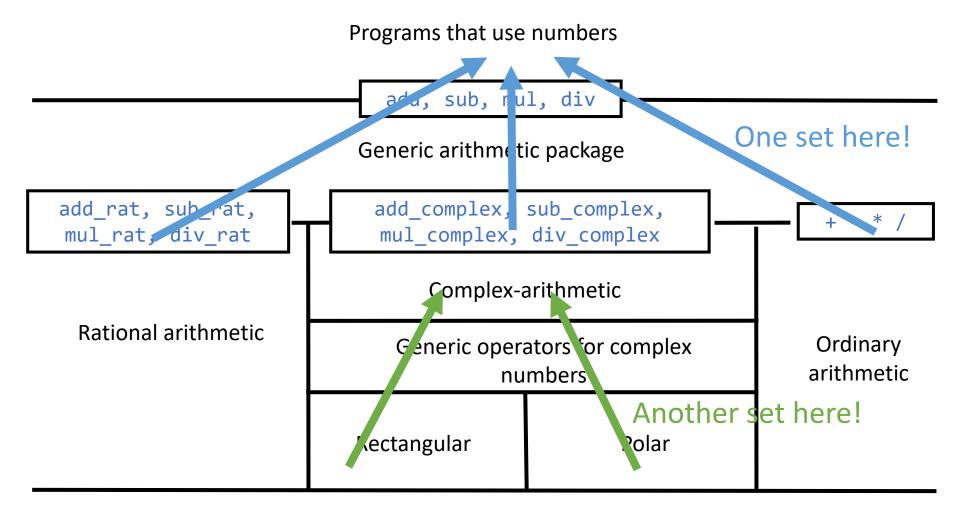
```
# interface to rest of the system
  def tag(z):
         return attach tag('complex', z)
  put('add', ('complex', 'complex'),
       lambda z1, z2: tag(add_complex(z1, z2)))
  put('sub', ('complex', 'complex'),
       lambda z1, z2: tag(sub complex(z1, z2)))
  put('mul', ('complex', 'complex'),
       lambda z1, z2: tag(mul_complex(z1, z2)))
  put('div', ('complex', 'complex'),
       lambda z1, z2: tag(div complex(z1, z2)))
  put('make_from_real_imag', 'complex',
      lambda x, y: tag(make_from_real_imag(x, y)))
  put('make_from_mag_ang', 'complex',
      lambda x, y: tag(make from mag ang (x, y)))
return 'done'
```

Tags upon tags



- There are actually two tags
- Why two tags?

Generic Arithmetic



Lists and primitive machine arithmetic

Layering Abstraction

- This is an IMPORTANT(!) idea:
 - As data passes down a layer, a tag is added
 - As data passes up a layer, the tag is stripped away
- Applicable to other areas of computer science and not just data abstraction
- Computer Networking!

Different data types

- So far, we are adding numbers of the same type: complex + complex.
- What happens if we need to add complex to rational?
- Answer: coercion
- Force (or convert) one data type into another.
- Insight: Exploit the hierarchy of numbers.

Number Hierarchy

Complex Real Rational Integer

 Thus we can coerce a number of a lower type to one of a higher type.

- But the reverse may not always be possible!
 - e.g. 3+4i to integer?

All seems like a blur?

Missions 11 and 12 will fix that.

Strategy #3: Message passing

- Previous strategies viewed functions as "intelligent":
 - They dispatch according to type of data.
- In message passing, it is the data that is "intelligent":
 - They know how to act on themselves.
 - You just "tell" the data what you want.

Message passing

```
def make from real imag(x, y):
    def dispatch(op):
      if op == 'real part':
            return x
      elif op == 'imag part':
           return y
      elif op == 'magnitude':
           return math.hypot(x, y)
      elif op == 'angle':
           return math.atan(y / x)
      else:
           raise Exception("Unknown op --
                   make from real_imag" + op)
    return dispatch
```

Message passing

- "Data" is the function dispatch.
- dispatch accepts a "message" op, and performs the necessary action according to the op.

```
def real_part(z):
    return z('real_part')
def imag_part(z):
    return z('imag_part')
```

Message passing

- This idea of "passing a message" to the data and letting the data do the job is the basis of object-oriented programming.
 - e.g. Java, C++, Smalltalk
 - Data are objects.
 - Functions are actions that objects perform.
 - Objects "talk" to other objects by passing messages

Summary

- Challenges of managing Multiple Representations
 - Matching data types to operations.
 - Resolving name conflicts.
 - Managing changes/future modifications
- Abstraction layer between the underlying representations and the user

Summary

- You have now seen 3 strategies for creating generic operators:
 - Dispatching on type
 - Data-directed programming
 - Message passing

- What are the pros/cons of each?
 - How does each resolve name conflicts?
 - What happens when a new type/operation is added?